

# ***Differential Equations Notes***

## ***Based on Lectures and "An Introduction to ODEs"***

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*Not in University of Cambridge  
skipped some talks irrelevant to contents*

*E-mail:* [not telling you](#)

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# Basic Calculus

## 1 Differentiation

### 1.1 Definitions and methods

**Definition 1.1** (Derivative). The derivative of a function  $f(x)$  wrt its argument  $x$  is the function

$$\frac{df}{dx} = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}.$$

We define higher derivatives recursively by

$$\frac{d^n f}{dx^n} = \frac{d}{dx} \left( \frac{d^{n-1} f}{dx^{n-1}} \right).$$

For the derivative to exist, we need

$$\lim_{h \rightarrow 0-} \frac{f(x+h) - f(x)}{h} = \lim_{h \rightarrow 0+} \frac{f(x+h) - f(x)}{h}.$$

Rules for differentiation:

1. **Chain rule:**  $(f(g(x)))' = f'(g(x))g'(x)$ .
2. **Product rule:**  $(u \cdot v)' = u \cdot v' + u' \cdot v$ .
3. **Leibniz's rule:** generalisation of product rule.<sup>1</sup>

$$\frac{d^n}{dx^n}(u \cdot v) = \sum_{k=0}^n \binom{n}{k} u^{(k)} v^{(n-k)}.$$

<sup>1</sup> There are multiple ways to prove, e.g. by induction.

### 1.2 Order of magnitude

The goal is to compare the sizes of functions, in the vicinity of specific points.

**Definition 1.2** (Little and Big o). We say  $f(x) = o(g(x))$  as  $x \rightarrow x_0$  if  $\lim_{x \rightarrow x_0} \frac{f(x)}{g(x)} = 0$ . We say  $f(x) = O(g(x))$  as  $x \rightarrow x_0$  if  $\exists M, \delta > 0, |x - x_0| < \delta \Rightarrow |f(x)| \leq M |g(x)|$ . The infinite case is defined similarly.

To find the tangent line to  $f$  at  $x_0$ , note that

$$\begin{aligned} \frac{df}{dx} \Big|_{x=x_0} &= \frac{f(x_0+h) - f(x_0)}{h} + \frac{o(h)}{h} \quad \text{when } h \rightarrow 0 \\ \Rightarrow f(x_0+h) &= f(x_0) + \frac{df}{dx} \Big|_{x=x_0} h + o(h) \quad \text{when } h \rightarrow 0 \end{aligned}$$

### 1.3 Taylor's Theorem and L'Hopital's Theorem

We want to approximate a function  $f(x)$  with a polynomial of order  $n$ :

$$f(x) = \underbrace{a_0 + a_1 x + \cdots + a_n x^n}_{P_n(x)}.$$

Differentiating recursively we get

$$P_n(x) = f(x_0) + (x - x_0)f'(x_0) + \cdots + \frac{(x - x_0)^n}{n!}f^{(n)}(x_0). \quad (1.1)$$

Alternatively, we can write  $f(x) = P_n(x) + E_n$ , where  $E_n$  is called the *remainder/error*.

By generalisation of  $f(x + h) = f(x) + hf'(x) + o(h)$ ,  $h \rightarrow 0$ , we get

$$f(x + h) = f(x) + hf'(x) + \frac{h^2}{2}f''(x) + \cdots + \frac{h^n}{n!}f^{(n)}(x) + o(h^n). \quad (1.2)$$

By refining the range of  $o(h^n)$  we get

**Theorem 1.1 (Taylor).** If the first  $n + 1$  derivatives of  $f(x)$  exist, then

$$f(x + h) = f(x) + hf'(x) + \frac{h^2}{2}f''(x) + \cdots + \frac{h^n}{n!}f^{(n)}(x) + O(h^{n+1}).$$

Using this we can prove

**Theorem 1.2 (L'Hopital).** Let  $f$  and  $g$  be differentiable at  $x = x_0$  and

$$\lim_{x \rightarrow x_0} f(x) = f(x_0) = 0, \quad \lim_{x \rightarrow x_0} g(x) = g(x_0) = 0.$$

**Proof.** (Not rigorous) As  $x \rightarrow x_0$ ,

$$\begin{aligned} \frac{f(x)}{g(x)} &= \frac{f(x_0) + (x - x_0)f'(x_0) + o(x - x_0)}{g(x_0) + (x - x_0)g'(x_0) + o(x - x_0)} \\ &= \frac{(x - x_0)f'(x_0) + o(x - x_0)}{(x - x_0)g'(x_0) + o(x - x_0)} \\ &\rightarrow \frac{f'(x_0)}{g'(x_0)}. \end{aligned}$$

□

Note that it can be applied recursively.

## 2 Integration

### 2.1 Definition

All functions mentioned are assumed to be well-behaved.

We evaluate the area under the curve of  $f(x)$  by considering

$$\sum_{n=0}^{N-1} f(x_n)\Delta x$$

where  $\Delta x = \frac{b-a}{N}$  and  $x_n = a + n\Delta x$ .

**Theorem 2.1 (MVT).** For a continuous function  $f(x)$ :

$$\int_{x_n}^{x_{n+1}} f(x) dx = f(x_c)(x_{n+1} - x_n) \quad \text{for some } x_c \in (x_n, x_{n+1}).$$

Estimate  $f(x_c)$  as follows:

$$f(x_c) = f(x_n) + O(x_c - x_n) = f(x_n) + O(x_{n+1} - x_n).$$

Hence

$$\begin{aligned} \int_{x_n}^{x_{n+1}} f(x) \, dx &= f(x_c)(x_{n+1} - x_n) \\ &= [f(x_n) + O(x_{n+1} - x_n)](x_{n+1} - x_n) \\ &= \Delta x f(x_n) + O(\Delta x^2). \end{aligned}$$

Therefore the error  $\epsilon = O(\Delta x^2)$ . It follows that

$$\int_a^b f(x) \, dx = \lim_{\Delta x \rightarrow 0} \left\{ \left[ \sum_{n=0}^{N-1} f(x_n) \Delta x \right] + O(N \Delta x^2) \right\}.$$

Hence

**Definition 2.1** (Definite integral).  $\int_a^b f(x) \, dx = \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} f(x_n) \Delta x$

## 2.2 Fundamental Theorem of Calculus

**Theorem 2.2** (FTC). Let

$$F(x) = \int_a^x f(t) \, dt,$$

then

$$\frac{dF}{dx} = f(x).$$

**Proof.** From the definition of derivative:

$$\begin{aligned} \frac{dF}{dx} &= \lim_{h \rightarrow 0} \frac{1}{h} \left\{ \int_a^{x+h} f(t) \, dt - \int_a^x f(t) \, dt \right\} \\ &= \lim_{h \rightarrow 0} \frac{1}{h} \int_x^{x+h} f(t) \, dt \\ &= \lim_{h \rightarrow 0} \frac{1}{h} (f(x)h + O(h^2)) \\ &= f(x). \end{aligned}$$

□

**Corollary 2.3.**

$$\begin{aligned} \frac{d}{dx} \int_x^b f(t) \, dt &= -f(x). \\ \frac{d}{dx} \int_a^{g(x)} f(t) \, dt &= \frac{d}{dx} F(g(x)) = \frac{dF}{dg} \frac{dg}{dx} = f(g(x)) \frac{dg}{dx}. \end{aligned}$$

**Definition 2.2** (Indefinite integral).

$$\int f(x) \, dx = \int_{x_0}^x f(t) \, dt.$$

## 2.3 Techniques of Integration

skipped

## 3 Introduction to multivariable functions

Lecture 5.

### 3.1 Partial derivative

**Definition 3.1.** The *partial derivative* of  $f(x, y)$  wrt  $x$  is

$$\left. \frac{\partial f}{\partial x} \right|_y = \lim_{\delta x \rightarrow 0} \frac{f(x + \delta x, y) - f(x, y)}{\delta x}. \quad (3.1)$$

Similarly

$$\left. \frac{\partial f}{\partial y} \right|_x = \lim_{\delta y \rightarrow 0} \frac{f(x, y + \delta y) - f(x, y)}{\delta y}.$$

We can take them in any order to form *cross derivatives*.

Note that

$$\frac{\partial^2 f}{\partial x \partial y} = \frac{\partial}{\partial x} \left( \frac{\partial f}{\partial y} \right) = \frac{\partial}{\partial y} \left( \frac{\partial f}{\partial x} \right). \quad (3.2)$$

### 3.2 Multivariable chain rule

**Theorem 3.1.** For well-behaved functions, we have

$$df = \frac{\partial f}{\partial x} dx + \frac{\partial f}{\partial y} dy \quad (3.3)$$

**Proof.** Note that

$$\begin{aligned} \delta f &= f(x + \delta x, y + \delta y) - f(x + \delta x, y) + f(x + \delta x, y) - f(x, y) \\ &= f(x + \delta x, y) + \delta y \frac{\partial f}{\partial y}(x + \delta x, y) + o(\delta y) - f(x + \delta x, y) \\ &\quad + f(x, y) + \delta x \frac{\partial f}{\partial x}(x, y) + o(\delta x) - f(x, y) \\ &= \delta y \frac{\partial f}{\partial y}(x + \delta x, y) + \delta x \frac{\partial f}{\partial x}(x, y) + o(\delta x) + o(\delta y) \\ &= \delta y \left( \frac{\partial f}{\partial y}(x, y) + \delta x \frac{\partial}{\partial x} \left( \frac{\partial f}{\partial y}(x, y) \right) + o(\delta x) \right) + \delta x \frac{\partial f}{\partial x}(x, y) + o(\delta x) + o(\delta y) \\ &= \delta y \frac{\partial f}{\partial y}(x, y) + \delta x \frac{\partial f}{\partial x}(x, y) + \delta x \delta y \frac{\partial}{\partial x} \left( \frac{\partial f}{\partial y}(x, y) \right) + o(\delta x) + o(\delta y) + o(\delta x \delta y). \end{aligned}$$

Taking limit gives the result.  $\square$

**Remark.** For  $f(x(t), y(t))$ , we have

$$\frac{df}{dt} = \lim_{\delta x, \delta y, \delta t \rightarrow 0} \left[ \frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt} \right] = \frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt}. \quad (3.4)$$

And integral form:

$$\int df = \int \frac{\partial f}{\partial x} dx + \int \frac{\partial f}{\partial y} dy \quad (3.5)$$

In this case we need to specify the *path* of integral as there might be some priority

issues.

### 3.3 Applications of multivariable chain rule

#### 3.3.1 Change of variables

It is often useful to write a DE in a different coordinate system before solving it. Need to transform the derivatives into the new coordinate system.

**Example.** Change from cartesian coordinates to polar coordinates:  $x = r \cos \theta, y = r \sin \theta$ . Firstly, write

$$f = f(x(r, \theta), y(r, \theta)).$$

We have

$$\frac{\partial f}{\partial r} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial r} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial r} = \frac{\partial f}{\partial x} \cos \theta + \frac{\partial f}{\partial y} \sin \theta.$$

By regarding  $\frac{\partial f}{\partial r}$  as  $\frac{df}{dr}$  with  $\theta$  fixed, we get this result.

Similar for other partial derivatives.

#### 3.3.2 Implicit Differentiation

Consider  $f(x, y, z) = c, c \in \mathbb{R}$ .  $f$  describes a surface in 3d space.  $f(x, y, z) = c$  implicitly defines  $x(y, z), y(x, z), z(x, y)$ . However, we can find  $\frac{\partial z}{\partial x}$  here using implicit differentiation.

Consider  $f(x, y, z(x, y)) = c$ .

$$df = \frac{\partial f}{\partial x} dx + \frac{\partial f}{\partial y} dy + \frac{\partial f}{\partial z} dz.$$

Finding the partial derivative for  $x$ :

$$\begin{aligned} \frac{\partial f}{\partial x} \Big|_y &= \frac{\partial f}{\partial x} \Big|_{yz} \frac{\partial x}{\partial x} \Big|_y + \frac{\partial f}{\partial y} \Big|_{xz} \frac{\partial y}{\partial x} \Big|_y + \frac{\partial f}{\partial z} \Big|_{xy} \frac{\partial z}{\partial x} \Big|_y \\ &= \frac{\partial f}{\partial x} \Big|_{yz} + \frac{\partial f}{\partial y} \Big|_{xz} \frac{\partial y}{\partial x} \Big|_y + \frac{\partial f}{\partial z} \Big|_{xy} \frac{\partial z}{\partial x} \Big|_y. \\ \iff \frac{\partial f}{\partial x} \Big|_y &= \frac{\partial f}{\partial x} \Big|_{yz} + \frac{\partial f}{\partial z} \Big|_{xy} \frac{\partial z}{\partial x} \Big|_y \\ \iff 0 &= \frac{\partial f}{\partial x} \Big|_{yz} + \frac{\partial f}{\partial z} \Big|_{xy} \frac{\partial z}{\partial x} \Big|_y \\ \iff \boxed{\frac{\partial z}{\partial x} \Big|_y} &= - \frac{\partial f / \partial x \Big|_{yz}}{\partial f / \partial z \Big|_{xy}} \end{aligned}$$

Notice the subscripts are very important since they describes different functions

Since  $\frac{\partial y}{\partial x} \Big|_y = 0$

Since  $f = c$  along the surface  $z(x, y)$ .

Note that  $\frac{\partial f}{\partial x} \Big|_{yz} \neq 0$  in general.

**Remark.** Reciprocal rule still holds as long as the same variable(s) are held fixed. e.g.

$$\frac{\partial r}{\partial x} \Big|_y = \frac{1}{\frac{\partial x}{\partial r} \Big|_y} \quad \text{but} \quad \frac{\partial r}{\partial x} \Big|_y \neq \frac{1}{\frac{\partial x}{\partial r} \Big|_\theta}.$$

#### 3.3.3 Differentiation of an integral wrt its parameters

Consider a family of functions  $f(x; \alpha)$ , where  $\alpha$  is the parameter. Define

$$I(\alpha) = \int_{a(\alpha)}^{b(\alpha)} f(x; \alpha) dx.$$



$$\begin{aligned}
\frac{dI}{d\alpha} &= \lim_{\delta\alpha \rightarrow 0} \frac{I(\alpha + \delta\alpha) - I(\alpha)}{\delta\alpha} \\
&= \lim_{\delta\alpha \rightarrow 0} \frac{1}{\delta\alpha} \left[ \int_{a(\alpha+\delta\alpha)}^{b(\alpha+\delta\alpha)} f(x; \alpha + \delta\alpha) dx - \int_{a(\alpha)}^{b(\alpha)} f(x; \alpha) dx \right] \\
&= \lim_{\delta\alpha \rightarrow 0} \frac{1}{\delta\alpha} \left[ \int_{a(\alpha)}^{b(\alpha)} f(x; \alpha + \delta\alpha) - f(x; \alpha) dx - \int_{a(\alpha)}^{a(\alpha+\delta\alpha)} f(x; \alpha + \delta\alpha) dx + \int_{b(\alpha)}^{b(\alpha+\delta\alpha)} f(x; \alpha + \delta\alpha) dx \right] \\
&= \int_{a(\alpha)}^{b(\alpha)} \frac{\partial f}{\partial \alpha} dx - f(a; \alpha) \lim_{\delta\alpha \rightarrow 0} \frac{a(\alpha + \delta\alpha) - a(\alpha)}{\delta\alpha} + f(b; \alpha) \lim_{\delta\alpha \rightarrow 0} \frac{b(\alpha + \delta\alpha) - b(\alpha)}{\delta\alpha}.
\end{aligned}$$

Draw a graph to understand the steps.

When  $\delta\alpha$  is very small, we can approximate the latter two integrals with the area of the rectangle of height  $f(a; \alpha)$  and width  $a(\alpha + \delta\alpha) - a(\alpha)$ .

Hence,

$$\frac{dI}{d\alpha} = \frac{d}{d\alpha} \int_{a(\alpha)}^{b(\alpha)} f(x; \alpha) dx = \int_{a(\alpha)}^{b(\alpha)} \frac{\partial f}{\partial \alpha} dx + f(b; \alpha) \frac{db}{d\alpha} - f(a; \alpha) \frac{da}{d\alpha}.$$

# First order linear ODEs

## 4 Terminology

**Definition 4.1.** An *ordinary differential equation* is a differential equation involving a function of one variable. A *partial differential equation* is (a) differential equation(s) involving a function of more than one variable.

$n$ th order DE: the highest order of derivative is  $n$ .

Linear: dependent variable appears linearly.

## 5 Prelude: Exponential functions

Consider  $f = a^x, a > 0$ , we have

$$\begin{aligned}
\frac{df}{dx} &= \lim_{h \rightarrow 0} \frac{a^{x+h} - a^x}{h} = a^x \lim_{h \rightarrow 0} \frac{a^h - 1}{h} \\
&= \lambda a^x.
\end{aligned}$$

Hence

**Definition 5.1.** Define  $\exp(x) = e^x$  as the solution to the DE

$$\frac{df}{dx} = f(x), \quad f(0) = 1.$$

Therefore  $e$  is the value of  $a$  such that  $\lambda = 1$ . i.e.,

$$\lim_{h \rightarrow 0} \frac{e^h - 1}{h} = 1.$$

Define  $\ln(x)$  as the inverse of  $e^x$  such that  $e^{\ln(x)} = x$ .

Consider  $a^x = e^{\ln(a)x}$ , so

$$\frac{df}{dx} = (\ln a) a^x, \quad \lambda = \ln a.$$

The exponential function is the *eigenfunction* of the differential operator.

The *eigenfunction* of an operator is unchanged by the action of the operator, except for a multiplicative scaling by the eigenvalue.

## 6 Rules for linear ODEs

1. Any linear homogeneous ODE with constant coefficients has solutions of form  $e^{\lambda x}$ , the eigenfunction. By *homogeneous* we mean that all terms involve the dependent variable or its derivatives.

This means that  $y = 0$  is a trivial solution for all homogeneous ODEs.

Constant coefficients imply that the independent variable does not appear explicitly in DE.

2. For linear homogeneous ODEs, any constant multiple of a solution is also a solution.
3. An  $n$ th order ODE has  $n$  independent solutions.

For constant coefficient ODEs, this rule follows from the fundamental theorem of algebra.

4. An  $n$ th order ODE requires  $n$  initial/boundary conditions.

## 7 Inhomogeneous(forced) first order ODEs with constant coefficients

### 7.1 Constant forcing

**Example.** Consider the equation

$$5y' - 3y = 10.$$

Solution steps:

1. Write the general solution  $y = y_p + y_c$  where  $y_p$  is a *particular integral* and  $y_c$  is a complementary function
2. Find  $y_p$  by simply setting  $y' = 0$ . In this case,  $y = -10/3$ .
3. Insert general solution into DE:

$$\begin{aligned} 5(y_p + y_c)' - 3(y_p + y_c) &= 10 \\ \iff 5y_c' + 10 - 3y_c &= 10 \\ \iff 5y_c - 3y_c' &= 0. \end{aligned}$$

Note that  $y_c$  is a solution to corresponding homogeneous equation.

4. Solve for  $y_c$ . In this case,  $y_c = Ae^{3x/5}$ .
5. Combine  $y_p$  and  $y_c$ .

## 7.2 Eigenfunction forcing

Example problem: In a sample of rock, isotope A decays to isotope B at a rate proportional to  $a$ , the number of nuclei of A. B decays to C at a rate proportional to  $b$ , the number of nuclei of B. Find  $b(t)$ .

We have

$$\begin{aligned}\frac{da}{dt} &= -k_a a \implies a = a_0 e^{-k_a t} \\ \frac{db}{dt} &= k_a a - k_b b,\end{aligned}$$

which means  $\dot{b} + k_b b = k_a a_0 e^{-k_a t}$ . RHS is called a *forcing term*, and it is an eigenfunction of differential operator.

We *guess* the form of the particular integral

$$b_p = c e^{-k_a t},$$

then the equation becomes

$$-k_a c + k_b c = k_a a_0 \iff c = \frac{k_a}{k_b - k_a} a_0, \quad \text{for } k_b \neq k_a.$$

Since the general solution for the DE is  $b = b_p + b_c$ ,

$$\dot{b}_c + k_b b_c = 0 \iff b_c = D e^{-k_b t}.$$

Hence

$$b = \frac{k_a}{k_b - k_a} a_0 e^{-k_a t} + D e^{-k_b t}.$$

If  $b(0) = 0$ ,  $D = -c$ , then

$$b = \frac{k_a}{k_b - k_a} a_0 (e^{-k_a t} - e^{-k_b t}).$$

Taking the ratio of  $b$  and  $a$ :

$$\frac{b(t)}{a(t)} = \frac{k_a}{k_b - k_a} (1 - e^{(k_a - k_b)t}).$$

We can date the age without knowing  $a_0$  is. This result allows rocks and other materials to be dated by measuring ratio of isotopes.

## 8 First order ODEs of non-constant coefficients

The general form is

$$a(x)y' + b(x)y = c(x).$$

The standard form is

$$y' + p(x)y = f(x).$$

Solved using *integrating factors*, multiply by IF  $\mu$ :

$$\mu y' + (\mu p)y = \mu f.$$

If  $\mu p = \mu'$ , LHS =  $(\mu y)'$  by product rule. Hence we want  $p = \mu'/\mu$ .

$$\int p \, dx = \int \frac{\mu'}{\mu} \, dx = \ln \mu \implies \boxed{\mu = e^{\int p(x) \, dx}}.$$

Thus the DE becomes

$$(\mu y)' = \mu f \iff y = \frac{1}{\mu} \int \mu f \, dx.$$

## 9 Discrete equations

A *discrete equation* is an equation involving a function evaluated at a discrete set of points. Lecture 8

### 9.1 Numerical integration

Consider a discrete representation of  $y(x)$ ,  $y(x_1), \dots, y(x_n)$ . One approximation to  $y'$  is

$$\left. \frac{dy}{dx} \right|_{x_n} \approx \frac{y_{n+1} - y_n}{h}, \quad h = \frac{x_n}{n},$$

given that  $x_i$  are uniformly distributed. This is called the *Forward Euler* approximation, but it is not the best approximation of the derivative in most contexts.

**Example.** Consider  $5y' - 3y = 0$ . We can approximate the equation by

$$5 \frac{y_{n+1} - y_n}{h} - 3y = 0,$$

which is called a *difference equation*, and deduce that

$$y_{n+1} = \left(1 + \frac{3h}{5}\right) y_n,$$

which is called a *recurrence relation*.

Apply recurrence relation repeatedly:

$$y_n = \left(1 + \frac{3h}{5}\right)^n y_0 = \left(1 + \frac{3x_n}{5n}\right)^n y_0.$$

Euler's definition of  $e^x$  is

$$e^x = \lim_{n \rightarrow \infty} \left(1 + \frac{x}{n}\right)^n.$$

It can be shown that this definition is equivalent to the previous definition. Hence

$$y(x) = \lim_{n \rightarrow \infty} y_n = y_0 e^{3x/5}.$$

Note for finite  $n$ ,  $y_n < y(x)$ .

### 9.2 Series solutions

A powerful way to solve ODEs is to seek solutions in the form of an infinite power series.

$$y(x) = \sum_{n=0}^{\infty} a_n x^n.$$

Plug into DE and find a solution.

**Example.** Consider  $5y' - 3y = 0$ . Let

$$y(x) = \sum_{n=0}^{\infty} a_n x^n, \quad y' = \sum_{n=1}^{\infty} n a_n x^{n-1}.$$

Multiply both sides by  $x$ :

$$\begin{aligned} xy' &= \sum_{n=1}^{\infty} n a_n x^n, \\ xy &= \sum_{n=0}^{\infty} a_n x^{n+1} = \sum_{n=1}^{\infty} a_{n-1} x^n. \end{aligned}$$

Then the DE becomes

$$\begin{aligned} 5 \sum_{n=1}^{\infty} n a_n x^n - 3 \sum_{n=1}^{\infty} a_{n-1} x^n &= 0 \\ \iff \sum_{n=1}^{\infty} x^n (5n a_n - 3a_{n-1}) &= 0. \end{aligned}$$

This holds for every  $x \in \mathbb{R}$ , so it holds if and only if

$$\forall x \in \mathbb{R}, 5n a_n - 3a_{n-1} = 0 \iff a_n = \frac{3}{5n} a_{n-1} \iff a_n = \left(\frac{3}{5}\right)^n \frac{a_0}{n!}.$$

Hence

$$y = a_0 \sum_{n=0}^{\infty} \left(\frac{3}{5}\right)^n \frac{x^n}{n!} = a_0 e^{3x/5}.$$

This converges for all  $x$ , so  $y(x) = a_0 e^{3x/5}$  is a solution.

# First order nonlinear ODEs

General form is

$$Q(x, y) \frac{dy}{dx} + P(x, y) = 0. \quad (9.1)$$

## 10 Separable equations

(10.1) is separable if and only if it can be written in the form

$$q(y)dy = p(x)dx,$$

and we simply solve  $x, y$  by integrating both sides.

## 11 Exact equations

(10.1) is an *exact equation* if and only if

$$Q(x, y)dy + P(x, y)dx \quad (*)$$

is an *exact differential* of function  $f(x, y)$ . i.e.,  $df = Qdy + Pdx$ . If this holds, then (10.1) implies that  $df = 0$  and  $f(x, y)$  is constant. We can use multivariable chain rule to check.

$$df = \frac{\partial f}{\partial x}dx + \frac{\partial f}{\partial y}dy \implies \frac{\partial f}{\partial x} + \frac{\partial f}{\partial y} \frac{dy}{dx} = 0.$$

Comparing with (10.1), if (\*) is an exact differential, then  $\exists f(x)$  such that

$$\frac{\partial f}{\partial x} = P(x, y), \quad \frac{\partial f}{\partial y} = Q(x, y). \quad (**)$$

Hence

$$\frac{\partial^2 f}{\partial y \partial x} = \frac{\partial P}{\partial y} \wedge \frac{\partial^2 f}{\partial x \partial y} = \frac{\partial Q}{\partial x}$$

$$\iff \boxed{\frac{\partial P}{\partial y} = \frac{\partial Q}{\partial x}}.$$

If it holds throughout a *simply connected* domain  $\mathcal{D}$ , then  $Pdx + Qdy$  is an exact differential of a single-valued function  $f(x, y)$  in  $D$ . Hence we can use this to check exact equations.

$f(x, y)$  can be found by integrating (\*\*).

**Example.** Consider

$$6y(y - x) \frac{dy}{dx} + (2x - 3y^2) = 0.$$

Here  $P = 2x - 3y^2$ ,  $Q = 6y(y - x)$ . We have

$$\frac{\partial P}{\partial y} = -6y = \frac{\partial Q}{\partial x},$$

so it is an exact equation. Note that

$$\int \frac{\partial f}{\partial x} dx = x^2 - 3xy^2 + h(y),$$

$$\frac{\partial f}{\partial y} = (x^2 - 3xy^2 + h(y))'_y = -6xy + h'(y) = 6y(y - x),$$

$$\implies h' = 6y^2 \implies h = 2y^3.$$

Hence  $f(x, y) = x^2 - 3xy^2 + 2y^3 + C$  and

$$x^2 - 3xy^2 + 2y^3 = C$$

is the general solution.

## 12 Isoclines and solution curves

Nonlinear equations are not guaranteed to have simple/closed form solutions. Nevertheless we can analyze the behaviour of the system without solving. Lecture 9.

Consider an ODE of the form

$$\frac{dy}{dt} = f(y, t).$$

Each initial condition will give a different solution curve.

**Example.** Consider the equation

$$\frac{dy}{dt} = t(1 - y^2) = f(y, t). \quad (*)$$

It is separable:

$$\int \frac{dy}{1-y^2} = \int t dt$$

$$\Rightarrow y = \frac{A - e^{-t^2}}{A + e^{-t^2}}.$$

This general solution produces a family of solution curves, parameterised by  $A$ .

**Definition 12.1** (Isocline). An *isocline* is the curve along which  $f = \dot{y} = C$ , where  $C$  is a constant.

Procedure of drawing a curve: draw isoclines, inspect the slope of  $y$ , draw a vector field, and plot the lines.

**Remark.** Since  $f(y, t)$  is single-valued, any two solution curves do not cross.

## 13 Fixed(equilibrium) points

**Definition 13.1.** A fixed point is a point where

$$\frac{dy}{dt} = f(y, t) = 0.$$

A fixed point is called *stable*(*unstable*) if solution curves in a small neighbourhood of the fixed point converge(diverge) to(away) the fixed point.

We can analyze the stability of fixed points using a *perturbation* analysis.

Let  $y = a$  be a fixed point of  $\frac{dy}{dt} = f(y, t)$ , i.e.  $f(a, t) = 0$ . Consider a small perturbation from the fixed point:  $y = a + \epsilon(t)$ . We have

$$\begin{aligned} \frac{d\epsilon}{dt} &= \frac{d(y - a)}{dt} = \frac{dy}{dt} \\ &= f(a + \epsilon, t) \\ &= f(a, t) + \epsilon \frac{\partial f}{\partial y}(a, t) + O(\epsilon^2). \end{aligned}$$

For small  $\epsilon$ , we have

$$\frac{d\epsilon}{dt} \approx \epsilon \frac{\partial f}{\partial y}(a, t).$$

Hence we've converted the non-linear ODE into a linear one wrt  $\epsilon$ .

If  $\lim_{t \rightarrow \infty} \epsilon = 0$ , then  $a$  is a stable fixed point. Conversely if  $\lim_{t \rightarrow \infty} \epsilon = \infty$ , then  $a$  is an unstable fixed point. If  $f'_y(a, t) = 0$ , then we need higher order terms in Taylor series.

**Example.** Consider  $f(y, t) = t(1 - y^2)$ . The fixed points are  $y = \pm 1$ .  $f'_y = -2yt$ . At  $y = 1$ , we have

$$\dot{\epsilon} \approx -2\epsilon t \Rightarrow \epsilon = \epsilon_0 e^{-t^2} \rightarrow 0.$$

Hence 1 is stable.

At  $y = -1$ ,  $\dot{\epsilon} = 2\epsilon t \Rightarrow \epsilon = \epsilon_0 e^{t^2} \rightarrow \infty$ . Hence  $-1$  is unstable.

## 14 Autonomous DEs

**Definition 14.1.** An *autonomous DE* is a special case when  $\dot{y} = f(y)$ .

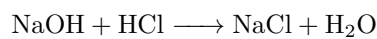
In this case, near fixed points  $y = a$ , we have  $\dot{\epsilon} = f'_y(a)\epsilon = \epsilon k$ , where  $k$  is constant. Hence  $\epsilon = \epsilon_0 e^{kt}$ . Therefore for autonomous DEs we have

$$\text{if } \begin{cases} f'(a) < 0 \Rightarrow & \text{stable F.P.} \\ f'(a) > 0 \Rightarrow & \text{unstable F.P.} \end{cases}$$

## 15 Phase Portraits

Another way to analyze solutions to a DE is using a geometrical representation of the solution called a *phase portrait*. Lecture 10

**Example** (Chemical kinetics). Consider the reaction



with

	NaOH	+	HCl	→	H <sub>2</sub> O	+	NaCl
Number of molecules	$a$		$b$		$c$		$c$
Initial number of molecules	$a_0$		$b_0$		0		0

A model of reaction rate is

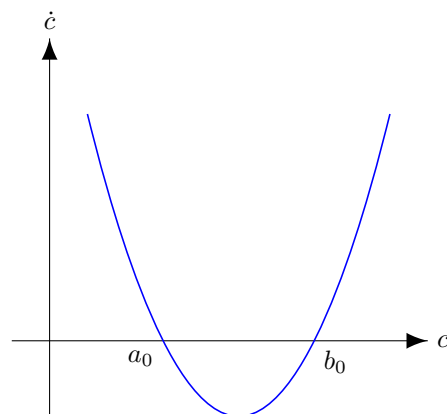
$$\frac{dc}{dt} = \lambda ab,$$

where  $\lambda$  is constant. Atoms are conserved:  $a = a_0 - c, b = b_0 - c$ . Then the equation can be written as

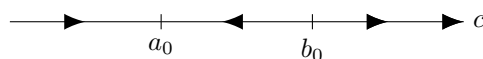
$$\frac{dc}{dt} = \lambda(a_0 - c)(b_0 - c).$$

This is an example of nonlinear first order ODE.

Plot a 2D phase portrait for this DE. One way is to plot  $\frac{dc}{dt}$  against  $t$ :



We can analyze the behaviour using 1D phase portrait:



arrows are drawn by sign of  $\dot{c}$ .

**Example** (Population dynamics). Let  $y(t)$  be population,  $\alpha y$  be birth rate and  $\beta y$  be death rate.



(a) Linear model

$$\frac{dy}{dt} = \alpha y - \beta y \Rightarrow y = y_0 e^{(\alpha-\beta)t}.$$

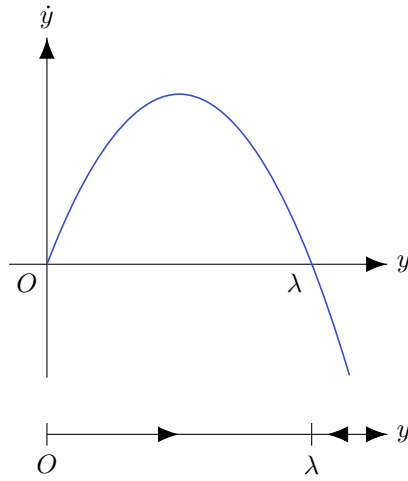
If  $\alpha > \beta$ , then  $\lim_{t \rightarrow \infty} y = \infty$ .

(b) Nonlinear model

$$\frac{dy}{dt} = (\alpha - \beta)y - \gamma y^2.$$

$\gamma y^2$  is dominant when  $y$  is large. It models increased death rate at high population.

Equivalently we have  $\dot{y} = ry(1 - \frac{y}{\lambda})$  where  $r = \alpha - \beta, \lambda = \frac{\alpha-\beta}{\gamma}$ .  $\lambda$  is called the *carrying capacity*.



## 16 Fixed points in discrete equations

### 16.1 Definitions

Consider a first order discrete(difference) equation of the form

$$x_{n+1} = f(x_n).$$

Define the *fixed point* as the value of  $x_n$  where  $x_{n+1} = x_n$ . That is, where

$$f(x_n) = x_n.$$

### 16.2 Stability

Use perturbation analysis to study its stability. Let  $x_f$  be a stable point of  $x_n$ , and perturb by a small  $\epsilon$ :

$$\begin{aligned} f(x_f + \epsilon) &= f(x_f) + \epsilon \left. \frac{df}{dx} \right|_{x_f} + O(\epsilon^2) \\ &= x_f + \epsilon \left. \frac{df}{dx} \right|_{x_f} + O(\epsilon^2) \\ &\approx x_f + \epsilon \left. \frac{df}{dx} \right|_{x_f}. \end{aligned}$$

Let  $x_n = x_f + \epsilon$ , then

$$x_{n+1} \approx x_f + \epsilon \left. \frac{df}{dx} \right|_{x_f}.$$

Therefore,

$$x_f \text{ is } \begin{cases} \text{stable} & \text{if } \left| \frac{df}{dx} \Big|_{x_f} \right| < 1 \\ \text{unstable} & \text{if } \left| \frac{df}{dx} \Big|_{x_f} \right| > 1 \end{cases}$$

### 16.3 Example: Logistic map

Nonlinear discrete population model

$$\frac{x_{n+1} - x_n}{\Delta t} = \lambda x_n - \gamma x_n^2.$$

Formally,

$$x_{n+1} = (\lambda \Delta t + 1)x_n - \gamma \Delta t x_n^2.$$

A simpler version is

$$x_{n+1} = rx_n(1 - x_n) = f(x_n).$$

This equation is the Logistic map.

**Fixed points:** Let  $f(x_n) = x_n$ . We have  $x_n(r - 1 - rx_n) = 0 \Rightarrow x_n = 0 \vee x_n = 1 - 1/r$ .

**Stability:** Write  $f(x) = rx(1 - x)$ , then

$$\frac{df}{dx} = r(1 - 2x).$$

Hence for  $x_n = 0$  we have  $\frac{df}{dx} \Big|_0 = r$ , so if  $0 < r < 1$ , it is stable. For  $r > 1$ , it is unstable.

For  $x_n = 1 - \frac{1}{r}$ ,  $\frac{df}{dx} \Big|_{x_n} = 2 - r$ . Then for  $0 < r < 1$  it is unphysical since  $x_n < 0$ . If  $1 < r < 3$ , it is stable. For  $r > 3$  it is unstable.